

DataSet_Analyiss

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Fifa17 dataset

<https://www.kaggle.com/artimous/complete-fifa-2017-player-dataset-global?select=FullData.csv>

Content:

- 17,000+ players
- 50+ attributes per player ranging from ball skills aggression etc.
- Player's attributes sourced from EA Sports' FIFA video game series, including the weekly updates
- Players from all around the globe
- URLs to their homepage
- Club logos
- Player images male and female
- National and club team data

Columns:

- 'Name'
- 'Nationality'
- 'National_Position'
- 'National_Kit'
- 'Club'
- 'Club_Position'
- 'Club_Kit'
- 'Club_Joining'
- 'Contract_Expiry'
- 'Rating'
- 'Height'
- 'Weight'
- 'Preferred_Foot'
- 'Birth_Date'
- 'Age'
- 'Preferred_Position'
- 'Work_Rate'
- 'Weak_foot'
- 'Skill_Moves'
- 'Ball_Control'
- 'Dribbling'
- 'Marking'
- 'Sliding_Tackle'
- 'Standing_Tackle'
- 'Aggression'
- 'Reactions'
- 'Attacking_Position'
- 'Interceptions'
- 'Vision'
- 'Composure'
- 'Crossing'
- 'Short_Pass'
- 'Long_Pass'
- 'Acceleration'
- 'Speed'
- 'Stamina'
- 'Strength'
- 'Balance'
- 'Agility'
- 'Jumping'
- 'Heading'
- 'Shot_Power'

- 'Finishing'
- 'Long_Shots'
- 'Curve'
- 'Freekick_Accuracy'
- 'Penalties'
- 'Volleys'
- 'GK_Positioning'
- 'GK_Diving'
- 'GK_Kicking'
- 'GK_Handling'
- 'GK_Reflexes'

Początkowo dane należało wyczyścić: m.in.: - dodać kolumny z skonwertowanymi danymi stringowymi na integer za pomocą techniki onehot encoding,

```
Fifa <- read.csv("Fifa17_ext.csv", header = TRUE, na.strings = "?")
Fifa <- subset(Fifa, select = -c(X))
dim(Fifa)
```

```
## [1] 17588    56
```

```
attach(Fifa)
head(Fifa)
```

##	Name	Nationality	Club	Club_Position	Club_Kit				
## 1	Cristiano Ronaldo	Portugal	Real Madrid	LW	7				
## 2	Lionel Messi	Argentina	FC Barcelona	RW	10				
## 3	Neymar	Brazil	FC Barcelona	LW	11				
## 4	Luis Suárez	Uruguay	FC Barcelona	ST	9				
## 5	Manuel Neuer	Germany	FC Bayern	GK	1				
## 6	De Gea	Spain	Manchester Utd	GK	1				
##	Club_Joining	Contract_Expiry	Rating	Height	Weight	Preffered_Foot	Birth_Date		
## 1	07/01/2009	2021	94	185	80	Right	02/05/1985		
## 2	07/01/2004	2018	93	170	72	Left	06/24/1987		
## 3	07/01/2013	2021	92	174	68	Right	02/05/1992		
## 4	07/11/2014	2021	92	182	85	Right	01/24/1987		
## 5	07/01/2011	2021	92	193	92	Right	03/27/1986		
## 6	07/01/2011	2019	90	193	82	Right	11/07/1990		
##	Age	Preffered_Position	Work_Rate	Weak_foot	Skill_Moves	Ball_Control			
## 1	32	LW/ST	High / Low	4	5	93			
## 2	29	RW	Medium / Medium	4	4	95			
## 3	25	LW	High / Medium	5	5	95			
## 4	30	ST	High / Medium	4	4	91			
## 5	31	GK	Medium / Medium	4	1	48			
## 6	26	GK	Medium / Medium	3	1	31			
##	Dribbling	Marking	Sliding_Tackle	Standing_Tackle	Aggression	Reactions			
## 1	92	22	23	31	63	96			
## 2	97	13	26	28	48	95			
## 3	96	21	33	24	56	88			
## 4	86	30	38	45	78	93			
## 5	30	10	11	10	29	85			
## 6	13	13	13	21	38	88			
##	Attacking_Position	Interceptions	Vision	Composure	Crossing	Short_Pass			
## 1	94	29	85	86	84	83			
## 2	93	22	90	94	77	88			
## 3	90	36	80	80	75	81			
## 4	92	41	84	83	77	83			
## 5	12	30	70	70	15	55			
## 6	12	30	68	60	17	31			
##	Long_Pass	Acceleration	Speed	Stamina	Strength	Balance	Agility	Jumping	Heading
## 1	77	91	92	92	80	63	90	95	85
## 2	87	92	87	74	59	95	90	68	71
## 3	75	93	90	79	49	82	96	61	62
## 4	64	88	77	89	76	60	86	69	77
## 5	59	58	61	44	83	35	52	78	25
## 6	32	56	56	25	64	43	57	67	21
##	Shot_Power	Finishing	Long_Shots	Curve	Freekick_Accuracy	Penalties	Volleys		
## 1	92	93	90	81	76	85	88		
## 2	85	95	88	89	90	74	85		
## 3	78	89	77	79	84	81	83		
## 4	87	94	86	86	84	85	88		
## 5	25	13	16	14	11	47	11		
## 6	31	13	12	21	19	40	13		
##	GK_Positioning	GK_Diving	GK_Kicking	GK_Handling	GK_Reflexes				
## 1	14	7	15	11	11				
## 2	14	6	15	11	8				
## 3	15	9	15	9	11				
## 4	33	27	31	25	37				
## 5	91	89	95	90	89				
## 6	86	88	87	85	90				
##	Nationality_encoded	Club_encoded	Preffered_Foot_encoded						
## 1	121	456	1						
## 2	5	200	0						
## 3	19	200	1						
## 4	154	200	1						
## 5	58	202	1						
## 6	137	358	1						
##	Preffered_Position_encoded	Work_Rate_encoded							
## 1	171	1							
## 2	236	8							
## 3	156	2							
## 4	265	2							
## 5	112	8							
## 6	112	8							

Regresja liniowa

Prosta regresja liniowa

Z użyciem prostej regresji liniowej sprawdzmy jaki wpływ ma wiek zawodnika na jego ogólną ocenę.

```
lmFitSimple <- lm(Rating ~ Age, data = Fifa)
summary(lmFitSimple)
```

```
##
## Call:
## lm(formula = Rating ~ Age, data = Fifa)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -36.105  -4.234  -0.234   3.927  26.153
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  48.50813    0.26258  184.74  <2e-16 ***
## Age          0.69355    0.01014   68.38  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 6.296 on 17586 degrees of freedom
## Multiple R-squared:  0.21, Adjusted R-squared:  0.21
## F-statistic: 4675 on 1 and 17586 DF, p-value: < 2.2e-16
```

Możemy sprawdzić poszczególne składowe modelu liniowego.

```
lmFitSimple$coefficients
```

```
## (Intercept)      Age
##  48.5081295    0.6935525
```

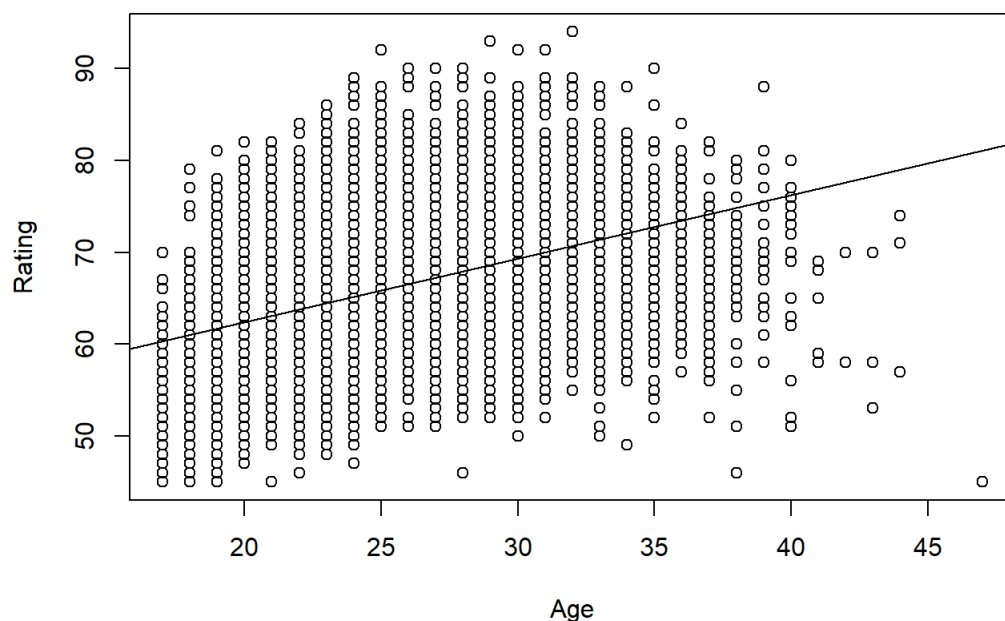
Następnie za pomocą funkcji predict() obliczamy przedziały ufności predykcji przewidujące średnie wartości

```
predict(lmFitSimple, data.frame(Age = c(5, 10, 15)), interval = "confidence")
```

```
##      fit      lwr      upr
## 1 51.97589 51.55860 52.39319
## 2 55.44365 55.12250 55.76481
## 3 58.91142 58.68358 59.13925
```

Prosta regresja liniowa na tle danych

```
plot(Age, Rating)
abline(lmFitSimple)
```



Regresja wielokrotna

```
lmFit.many <- lm(Rating ~ Nationality_encoded + Club_encoded + Club_Kit + Height + Weight + Preferred_Foot_encoded + Age + Preferred_Position_encoded)
summary(lmFit.many)
```

```
##
## Call:
## lm(formula = Rating ~ Nationality_encoded + Club_encoded + Club_Kit +
##     Height + Weight + Preferred_Foot_encoded + Age + Preferred_Position_encoded)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -35.539  -4.197  -0.242   3.890  26.287
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    52.0602600   1.5237360   34.166 < 2e-16 ***
## Nationality_encoded -0.0002375   0.0009915   -0.240  0.810706
## Club_encoded      0.0009958   0.0002586    3.850  0.000118 ***
## Club_Kit        -0.0304748   0.0025172  -12.107 < 2e-16 ***
## Height         -0.0382841   0.0110374   -3.469  0.000524 ***
## Weight          0.0699714   0.0108771    6.433 1.28e-10 ***
## Preferred_Foot_encoded -0.6478522   0.1123011  -5.769 8.11e-09 ***
## Age             0.6496982   0.0106364   61.082 < 2e-16 ***
## Preferred_Position_encoded 0.0006514   0.0005685    1.146 0.251899
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 6.252 on 17578 degrees of freedom
## (1 observation deleted due to missingness)
## Multiple R-squared:  0.221, Adjusted R-squared:  0.2206
## F-statistic: 623.2 on 8 and 17578 DF, p-value: < 2.2e-16
```

Nieliniowe transformacje predyktorów

```
lmFit5 <- lm(Rating ~ poly(Age, 5))
summary(lmFit5)
```

```
##
## Call:
## lm(formula = Rating ~ poly(Age, 5))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -23.3309  -3.8664  -0.4751   3.5722  24.4661
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    66.16619    0.04498 1470.929 < 2e-16 ***
## poly(Age, 5)1   430.46835    5.96558   72.159 < 2e-16 ***
## poly(Age, 5)2  -261.43378    5.96558  -43.824 < 2e-16 ***
## poly(Age, 5)3    47.07163    5.96558    7.891 3.18e-15 ***
## poly(Age, 5)4   -9.82544    5.96558   -1.647  0.0996 .
## poly(Age, 5)5  -25.49126    5.96558   -4.273 1.94e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 5.966 on 17582 degrees of freedom
## Multiple R-squared:  0.2908, Adjusted R-squared:  0.2906
## F-statistic: 1442 on 5 and 17582 DF, p-value: < 2.2e-16
```

```
anova(lmFitSimple, lmFit5)
```

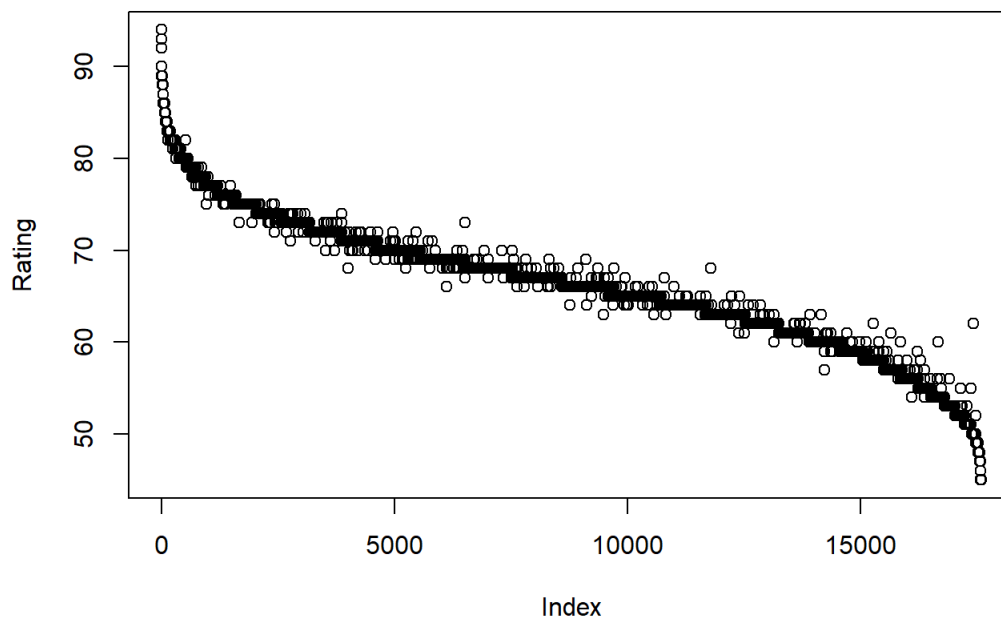
```
## Analysis of Variance Table
##
## Model 1: Rating ~ Age
## Model 2: Rating ~ poly(Age, 5)
##   Res.Df    RSS Df Sum of Sq    F    Pr(>F)
## 1  17586 697020
## 2  17582 625711   4     71310 500.94 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Klasyfikacja

TODO:

Bardziej skomplikowaną klasyfikację

```
plot(Rating)
```



```
fit.logistic <- glm(Preffered_Foot_encoded ~ Preffered_Position_encoded + Work_Rate_encoded + Nationality_encoded + Volleys + Penalties + Freekick_Accuracy, family = binomial, data = Fifa)
summary(fit.logistic)
```

```
##
## Call:
## glm(formula = Preffered_Foot_encoded ~ Preffered_Position_encoded +
##     Work_Rate_encoded + Nationality_encoded + Volleys + Penalties +
##     Freekick_Accuracy, family = binomial, data = Fifa)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.2616   0.4938   0.6384   0.7536   1.2470
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      1.4340759   0.0916233  15.652 < 2e-16 ***
## Preffered_Position_encoded 0.0009535   0.0002447   3.896 9.78e-05 ***
## Work_Rate_encoded      0.0206624   0.0067721   3.051 0.00228 **
## Nationality_encoded    0.0010190   0.0003799   2.682 0.00731 **
## Volleys            0.0113252   0.0019479   5.814 6.10e-09 ***
## Penalties          0.0091009   0.0021529   4.227 2.37e-05 ***
## Freekick_Accuracy    -0.0334104   0.0017073 -19.569 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 19087  on 17587  degrees of freedom
## Residual deviance: 18506  on 17581  degrees of freedom
## AIC: 18520
##
## Number of Fisher Scoring iterations: 4
```

```
probs.logistic <- predict(fit.logistic, type = "response")
head(probs.logistic)
```

```
##           1           2           3           4           5           6
## 0.7255330 0.6126944 0.6257871 0.7002957 0.8754540 0.8484157
```

```
pred.logistic <- ifelse(probs.logistic > 0.5, 1, 0)
conf.logistic <- table(pred.logistic, Preferred_Foot_encoded)
conf.logistic
```

```
##           Preferred_Foot_encoded
## pred.logistic    0    1
##           0    11    14
##           1  4083 13480
```

```
(conf.logistic[1, 2] + conf.logistic[2, 1]) / sum(conf.logistic)
```

```
## [1] 0.2329429
```

```
mean(pred.logistic != Preferred_Foot_encoded)
```

```
## [1] 0.2329429
```

```
n = dim(Fifa)[1]
train = 1:n/2
Fifa.test = Fifa[(n/2+1):n,]
Preferred_Foot_encoded.test = Preferred_Foot_encoded[(n/2+1):n]
```

```
fit.logistic <- glm(Preferred_Foot_encoded ~ Preferred_Position_encoded + Work_Rate_encoded + Nationality_
_encoded + Volleys + Penalties + Freekick_Accuracy, family = binomial, data = Fifa, subset = train)
summary(fit.logistic)
```

```
##
## Call:
## glm(formula = Preferred_Foot_encoded ~ Preferred_Position_encoded +
##       Work_Rate_encoded + Nationality_encoded + Volleys + Penalties +
##       Freekick_Accuracy, family = binomial, data = Fifa, subset = train)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.2536   0.4388   0.6508   0.7875   1.2957
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    1.4120677   0.0896710   15.747 < 2e-16 ***
## Preferred_Position_encoded 0.0008073   0.0002385    3.384 0.000713 ***
## Work_Rate_encoded    0.0137356   0.0063518    2.162 0.030581 *
## Nationality_encoded    0.0010518   0.0003655    2.878 0.004001 **
## Volleys          0.0111503   0.0017520    6.364 1.96e-10 ***
## Penalties         0.0127002   0.0020212    6.284 3.31e-10 ***
## Freekick_Accuracy   -0.0347229   0.0016083  -21.590 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 19578  on 17586  degrees of freedom
## Residual deviance: 18945  on 17580  degrees of freedom
## AIC: 18959
##
## Number of Fisher Scoring iterations: 4
```

```
probs.logistic <- predict(fit.logistic, Fifa.test, type = "response")
pred.logistic <- ifelse(probs.logistic > 0.5, 1, 0)
table(pred.logistic, Preferred_Foot_encoded.test)
```

```
##           Preferred_Foot_encoded.test
## pred.logistic    0    1
##           0    2    2
##           1 1939 6851
```



```
fit.logistic <- glm(PREFERRED_Foot_encoded ~ Volleys + Penalties + Freekick_Accuracy, family = binomial,
data = Fifa, subset = train)
summary(fit.logistic)
```

```
##
## Call:
## glm(formula = PREFERRED_Foot_encoded ~ Volleys + Penalties +
##     Freekick_Accuracy, family = binomial, data = Fifa, subset = train)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.2529   0.4446   0.6518   0.7890   1.3170
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    1.647697   0.065406  25.192 < 2e-16 ***
## Volleys         0.012381   0.001670   7.414 1.23e-13 ***
## Penalties       0.013174   0.002005   6.570 5.03e-11 ***
## Freekick_Accuracy -0.036150   0.001544 -23.407 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 19578  on 17586  degrees of freedom
## Residual deviance: 18968  on 17583  degrees of freedom
## AIC: 18976
##
## Number of Fisher Scoring iterations: 4
```

```
probs.logistic <- predict(fit.logistic, Fifa.test, type = "response")
pred.logistic <- ifelse(probs.logistic > 0.5, 1, 0)
table(pred.logistic, PREFERRED_Foot_encoded.test)
```

```
##              PREFERRED_Foot_encoded.test
## pred.logistic    0    1
##              0    2    5
##              1 1939 6848
```

LDA

```
fit.lda <- lda(PREFERRED_Foot_encoded ~ Volleys + Penalties + Freekick_Accuracy, data = Fifa, subset = train)
fit.lda
```

```
## Call:
## lda(PREFERRED_Foot_encoded ~ Volleys + Penalties + Freekick_Accuracy,
##     data = Fifa, subset = train)
##
## Prior probabilities of groups:
##      0      1
## 0.2448399 0.7551601
##
## Group means:
##      Volleys Penalties Freekick_Accuracy
## 0 49.76173  54.71760      54.03623
## 1 48.34576  53.17491      47.72193
##
## Coefficients of linear discriminants:
##              LD1
## Volleys         0.02910998
## Penalties       0.02879298
## Freekick_Accuracy -0.08282276
```

```
pred.lda <- predict(fit.lda, Fifa.test)
table(pred.lda$class, PREFERRED_Foot_encoded.test)
```

```
##      Preferred_Foot_encoded.test
##      0      1
##    0      2      5
##    1 1939 6848
```

```
max(pred.lda$posterior[, 2])
```

```
## [1] 0.9273505
```

```
max(pred.lda$posterior[, 1])
```

```
## [1] 0.527549
```

QDA

```
fit.qda <- qda(Preferred_Foot_encoded ~ Volleys + Penalties + Freekick_Accuracy, data = Fifa, subset = train)
fit.qda
```

```
## Call:
## qda(Preferred_Foot_encoded ~ Volleys + Penalties + Freekick_Accuracy,
##      data = Fifa, subset = train)
##
## Prior probabilities of groups:
##      0      1
## 0.2448399 0.7551601
##
## Group means:
##      Volleys Penalties Freekick_Accuracy
## 0 49.76173  54.71760      54.03623
## 1 48.34576  53.17491      47.72193
```

```
pred.qda <- predict(fit.qda, Fifa.test)
table(pred.qda$class, Preferred_Foot_encoded.test)
```

```
##      Preferred_Foot_encoded.test
##      0      1
##    0      7      7
##    1 1934 6846
```

kNN

```
train.set <- Fifa[1:(n/2), c("Volleys", "Freekick_Accuracy")]
test.set <- Fifa[(n/2 + 1):n, c("Volleys", "Freekick_Accuracy")]
Preferred_Foot_encoded.train <- Preferred_Foot_encoded[1:(n/2)]
pred.knn.1 <- knn(train.set, test.set, Preferred_Foot_encoded.train, k = 1)
table(pred.knn.1, Preferred_Foot_encoded.test)
```

```
##      Preferred_Foot_encoded.test
## pred.knn.1      0      1
##      0  325  925
##      1 1616 5928
```

```
mean(pred.knn.1[1:8793] != Preferred_Foot_encoded.test[1:8793])
```

```
## [1] 0.2889799
```

Selekcja cech dla modeli liniowych

```
fit.bs <- regsubsets(Rating ~ . - (Name + Nationality + Club + Club_Position + Club_Joining + Birth_Date
+ Preferred_Foot + Preferred_Position + Work_Rate), data = Fifa, nvmax = 46)
summary(fit.bs)
```

```
## Subset selection object
## Call: regsubsets.formula(Rating ~ . - (Name + Nationality + Club +
##       Club_Position + Club_Joining + Birth_Date + Preferred_Foot +
##       Preferred_Position + Work_Rate), data = Fifa, nvmax = 46)
## 46 Variables (and intercept)
##
##               Forced in Forced out
## Club_Kit                FALSE      FALSE
## Contract_Expiry          FALSE      FALSE
## Height                   FALSE      FALSE
## Weight                   FALSE      FALSE
## Age                     FALSE      FALSE
## Weak_foot                FALSE      FALSE
## Skill_Moves              FALSE      FALSE
## Ball_Control             FALSE      FALSE
## Dribbling                FALSE      FALSE
## Marking                  FALSE      FALSE
## Sliding_Tackle           FALSE      FALSE
## Standing_Tackle          FALSE      FALSE
## Aggression               FALSE      FALSE
## Reactions                FALSE      FALSE
## Attacking_Position       FALSE      FALSE
## Interceptions            FALSE      FALSE
## Vision                   FALSE      FALSE
## Composure                FALSE      FALSE
## Crossing                 FALSE      FALSE
## Short_Pass               FALSE      FALSE
## Long_Pass                FALSE      FALSE
## Acceleration             FALSE      FALSE
## Speed                    FALSE      FALSE
## Stamina                  FALSE      FALSE
## Strength                 FALSE      FALSE
## Balance                  FALSE      FALSE
## Agility                  FALSE      FALSE
## Jumping                  FALSE      FALSE
## Heading                  FALSE      FALSE
## Shot_Power               FALSE      FALSE
## Finishing                FALSE      FALSE
## Long_Shots               FALSE      FALSE
## Curve                    FALSE      FALSE
## Freekick_Accuracy        FALSE      FALSE
## Penalties                FALSE      FALSE
## Volleys                  FALSE      FALSE
## GK_Positioning           FALSE      FALSE
## GK_Diving                FALSE      FALSE
## GK_Kicking               FALSE      FALSE
## GK_Handling              FALSE      FALSE
## GK_Reflexes              FALSE      FALSE
## Nationality_encoded       FALSE      FALSE
## Club_encoded              FALSE      FALSE
## Preferred_Foot_encoded    FALSE      FALSE
## Preferred_Position_encoded FALSE      FALSE
## Work_Rate_encoded         FALSE      FALSE
## 1 subsets of each size up to 46
## Selection Algorithm: exhaustive
##
##      Club_Kit Contract_Expiry Height Weight Age Weak_foot Skill_Moves
## 1  ( 1 )  " "      " "      " "      " "      " "      " "
## 2  ( 1 )  " "      " "      " "      " "      " "      " "
## 3  ( 1 )  " "      " "      " "      " "      " "      " "
## 4  ( 1 )  " "      " "      " "      " "      " "      " "
## 5  ( 1 )  " "      " "      " "      " "      " "      " "
## 6  ( 1 )  " "      " "      " "      " "      " "      " "
## 7  ( 1 )  " "      " "      " "      " "      " "      "★"
## 8  ( 1 )  " "      " "      " "      " "      " "      "★"
## 9  ( 1 )  " "      " "      " "      " "      " "      "★"
## 10 ( 1 )  " "      " "      " "      " "      " "      "★"
## 11 ( 1 )  " "      " "      " "      " "      " "      "★"
## 12 ( 1 )  " "      " "      " "      " "      " "      "★"
## 13 ( 1 )  " "      " "      " "      " "      " "      "★"
```

##	15	(1)	" "	" "	" "	" "	" "
##	14	(1)	" "	" "	" "	" "	" * "
##	15	(1)	" "	" "	" "	" "	" * "
##	16	(1)	" "	" "	" "	" "	" * "
##	17	(1)	" "	" "	" "	" "	" * "
##	18	(1)	" "	" "	" "	" "	" * "
##	19	(1)	" "	" "	" "	" "	" * "
##	20	(1)	" "	" "	" "	" "	" * "
##	21	(1)	" "	" "	" "	" "	" * "
##	22	(1)	" "	" "	" "	" "	" * "
##	23	(1)	" "	" "	" "	" "	" * "
##	24	(1)	" "	" "	" "	" "	" * "
##	25	(1)	" "	" "	" "	" "	" * "
##	26	(1)	" "	" "	" "	" "	" * "
##	27	(1)	" "	" * "	" "	" "	" * "
##	28	(1)	" * "	" * "	" "	" "	" * "
##	29	(1)	" * "	" * "	" "	" * "	" * "
##	30	(1)	" * "	" * "	" "	" * "	" * "
##	31	(1)	" * "	" * "	" "	" * "	" * "
##	32	(1)	" * "	" * "	" "	" * "	" * "
##	33	(1)	" * "	" * "	" "	" * "	" * "
##	34	(1)	" * "	" * "	" "	" * "	" * "
##	35	(1)	" * "	" * "	" "	" * "	" * "
##	36	(1)	" * "	" * "	" "	" * "	" * "
##	37	(1)	" * "	" * "	" "	" * "	" * "
##	38	(1)	" * "	" * "	" "	" * "	" * "
##	39	(1)	" * "	" * "	" "	" * "	" * "
##	40	(1)	" * "	" * "	" * "	" * "	" * "
##	41	(1)	" * "	" * "	" * "	" * "	" * "
##	42	(1)	" * "	" * "	" * "	" * "	" * "
##	43	(1)	" * "	" * "	" * "	" * "	" * "
##	44	(1)	" * "	" * "	" * "	" * "	" * "
##	45	(1)	" * "	" * "	" * "	" * "	" * "
##	46	(1)	" * "	" * "	" * "	" * "	" * "
##			Ball_Control	Dribbling	Marking	Sliding_Tackle	Standing_Tackle
##	1	(1)	" "	" "	" "	" "	" "
##	2	(1)	" "	" "	" "	" "	" "
##	3	(1)	" * "	" "	" "	" "	" "
##	4	(1)	" * "	" "	" "	" "	" "
##	5	(1)	" * "	" "	" "	" "	" * "
##	6	(1)	" * "	" "	" "	" "	" * "
##	7	(1)	" * "	" "	" "	" "	" * "
##	8	(1)	" * "	" "	" "	" "	" * "
##	9	(1)	" * "	" "	" "	" "	" * "
##	10	(1)	" * "	" "	" "	" "	" * "
##	11	(1)	" * "	" "	" "	" "	" * "
##	12	(1)	" * "	" "	" "	" "	" * "
##	13	(1)	" * "	" "	" "	" "	" * "
##	14	(1)	" * "	" "	" "	" "	" * "
##	15	(1)	" * "	" "	" "	" "	" * "
##	16	(1)	" * "	" "	" "	" "	" * "
##	17	(1)	" * "	" "	" "	" "	" * "
##	18	(1)	" * "	" "	" "	" "	" * "
##	19	(1)	" * "	" "	" "	" "	" * "
##	20	(1)	" * "	" "	" "	" "	" * "
##	21	(1)	" * "	" "	" "	" "	" * "
##	22	(1)	" * "	" "	" "	" "	" * "
##	23	(1)	" * "	" "	" "	" "	" * "
##	24	(1)	" * "	" "	" "	" "	" * "
##	25	(1)	" * "	" "	" "	" "	" * "
##	26	(1)	" * "	" "	" "	" "	" * "
##	27	(1)	" * "	" "	" "	" "	" * "
##	28	(1)	" * "	" "	" "	" "	" * "
##	29	(1)	" * "	" "	" "	" "	" * "
##	30	(1)	" * "	" "	" "	" "	" * "
##	31	(1)	" * "	" "	" "	" "	" * "
##	32	(1)	" * "	" * "	" "	" "	" * "
##	33	(1)	" * "	" * "	" "	" "	" * "
##	34	(1)	" * "	" * "	" "	" "	" * "
##	35	(1)	" * "	" * "	" "	" "	" * "
##	36	(1)	" * "	" * "	" * "	" "	" * "
##	37	(1)	" * "	" * "	" * "	" "	" * "
##	38	(1)	" * "	" * "	" * "	" "	" * "

##	39	(1)	##*	##*	##*	##*	##*
##	40	(1)	##*	##*	##*	##*	##*
##	41	(1)	##*	##*	##*	##*	##*
##	42	(1)	##*	##*	##*	##*	##*
##	43	(1)	##*	##*	##*	##*	##*
##	44	(1)	##*	##*	##*	##*	##*
##	45	(1)	##*	##*	##*	##*	##*
##	46	(1)	##*	##*	##*	##*	##*
##			Aggression Reactions Attacking_Position Interceptions Vision				
##	1	(1)	##	##*	##	##	##
##	2	(1)	##	##*	##	##	##
##	3	(1)	##	##*	##	##	##
##	4	(1)	##	##*	##	##	##
##	5	(1)	##	##*	##	##	##
##	6	(1)	##	##*	##	##	##
##	7	(1)	##	##*	##	##	##
##	8	(1)	##	##*	##	##	##
##	9	(1)	##	##*	##	##	##
##	10	(1)	##	##*	##	##	##
##	11	(1)	##	##*	##	##	##
##	12	(1)	##	##*	##	##	##
##	13	(1)	##	##*	##	##	##
##	14	(1)	##	##*	##*	##	##
##	15	(1)	##	##*	##*	##	##
##	16	(1)	##	##*	##*	##	##
##	17	(1)	##	##*	##*	##	##
##	18	(1)	##	##*	##*	##	##
##	19	(1)	##	##*	##*	##	##
##	20	(1)	##	##*	##*	##	##
##	21	(1)	##	##*	##*	##	##
##	22	(1)	##	##*	##*	##	##
##	23	(1)	##	##*	##*	##	##
##	24	(1)	##	##*	##*	##	##
##	25	(1)	##	##*	##*	##	##
##	26	(1)	##	##*	##*	##	##
##	27	(1)	##	##*	##*	##	##
##	28	(1)	##	##*	##*	##	##
##	29	(1)	##	##*	##*	##	##
##	30	(1)	##	##*	##*	##	##
##	31	(1)	##	##*	##*	##	##
##	32	(1)	##	##*	##*	##	##
##	33	(1)	##	##*	##*	##	##
##	34	(1)	##	##*	##*	##	##
##	35	(1)	##	##*	##*	##	##*
##	36	(1)	##	##*	##*	##	##
##	37	(1)	##	##*	##*	##	##*
##	38	(1)	##*	##*	##*	##	##*
##	39	(1)	##*	##*	##*	##	##*
##	40	(1)	##*	##*	##*	##	##*
##	41	(1)	##*	##*	##*	##*	##*
##	42	(1)	##*	##*	##*	##*	##*
##	43	(1)	##*	##*	##*	##*	##*
##	44	(1)	##*	##*	##*	##*	##*
##	45	(1)	##*	##*	##*	##*	##*
##	46	(1)	##*	##*	##*	##*	##*
##			Composure Crossing Short_Pass Long_Pass Acceleration Speed Stamina				
##	1	(1)	##	##	##	##	##
##	2	(1)	##*	##	##	##	##
##	3	(1)	##	##	##	##	##
##	4	(1)	##	##	##	##	##
##	5	(1)	##	##	##	##	##
##	6	(1)	##	##	##	##	##
##	7	(1)	##	##	##	##	##
##	8	(1)	##	##	##	##	##
##	9	(1)	##	##	##	##	##*
##	10	(1)	##*	##	##	##	##*
##	11	(1)	##*	##	##	##	##*
##	12	(1)	##*	##*	##	##	##*
##	13	(1)	##*	##*	##	##	##*
##	14	(1)	##*	##*	##	##	##*
##	15	(1)	##*	##*	##	##	##*
##	16	(1)	##*	##*	##	##	##*
##	17	(1)	##*	##*	##*	##*	##

##			Strength	Balance	Agility	Jumping	Heading	Shot_Power	Finishing
## 1	(1)		""	""	""	""	""	""	""
## 2	(1)		""	""	""	""	""	""	""
## 3	(1)		""	""	""	""	""	""	""
## 4	(1)		""	""	""	""	*"	""	""
## 5	(1)		""	""	""	""	*"	""	""
## 6	(1)		""	""	""	""	*"	""	""
## 7	(1)		""	""	""	""	*"	""	""
## 8	(1)		*"	""	""	""	*"	""	""
## 9	(1)		*"	""	""	""	*"	""	""
## 10	(1)		*"	""	""	""	*"	""	""
## 11	(1)		*"	""	""	""	*"	""	""
## 12	(1)		*"	""	""	""	*"	""	""
## 13	(1)		*"	""	""	""	*"	""	""
## 14	(1)		*"	""	""	""	*"	""	""
## 15	(1)		*"	""	""	""	*"	""	""
## 16	(1)		*"	""	""	""	*"	""	""
## 17	(1)		*"	""	""	""	*"	""	""
## 18	(1)		*"	*"	""	""	*"	""	""
## 19	(1)		*"	*"	""	""	*"	*"	""
## 20	(1)		*"	*"	""	*"	*"	*"	""
## 21	(1)		*"	*"	""	*"	*"	""	*"
## 22	(1)		*"	*"	""	*"	*"	""	*"
## 23	(1)		*"	*"	""	*"	*"	""	*"
## 24	(1)		*"	*"	""	*"	*"	*"	*"
## 25	(1)		*"	*"	""	*"	*"	*"	*"
## 26	(1)		*"	*"	""	*"	*"	*"	*"
## 27	(1)		*"	*"	""	*"	*"	*"	*"
## 28	(1)		*"	*"	""	*"	*"	*"	*"
## 29	(1)		*"	*"	""	*"	*"	*"	*"
## 30	(1)		*"	*"	""	*"	*"	*"	*"
## 31	(1)		*"	*"	""	*"	*"	*"	*"
## 32	(1)		*"	*"	""	*"	*"	*"	*"
## 33	(1)		*"	*"	""	*"	*"	*"	*"
## 34	(1)		*"	*"	""	*"	*"	*"	*"
## 35	(1)		*"	*"	""	*"	*"	*"	*"
## 36	(1)		*"	*"	""	*"	*"	*"	*"
## 37	(1)		*"	*"	""	*"	*"	*"	*"
## 38	(1)		*"	*"	""	*"	*"	*"	*"
## 39	(1)		*"	*"	""	*"	*"	*"	*"
## 40	(1)		*"	*"	""	*"	*"	*"	*"
## 41	(1)		*"	*"	""	*"	*"	*"	*"
## 42	(1)		*"	*"	""	*"	*"	*"	*"
## 43	(1)		*"	*"	""	*"	*"	*"	*"

##	43	(1)	##	##	##	##	##	##	##
##	44	(1)	##	##	##	##	##	##	##
##	45	(1)	##	##	##	##	##	##	##
##	46	(1)	##	##	##	##	##	##	##
##			Long_Shots	Curve	Freekick_Accuracy	Penalties	Volleys	GK_Positioning	
##	1	(1)	##	##	##	##	##	##	##
##	2	(1)	##	##	##	##	##	##	##
##	3	(1)	##	##	##	##	##	##	##
##	4	(1)	##	##	##	##	##	##	##
##	5	(1)	##	##	##	##	##	##	##
##	6	(1)	##	##	##	##	##	##	##
##	7	(1)	##	##	##	##	##	##	##
##	8	(1)	##	##	##	##	##	##	##
##	9	(1)	##	##	##	##	##	##	##
##	10	(1)	##	##	##	##	##	##	##
##	11	(1)	##	##	##	##	##	##	##
##	12	(1)	##	##	##	##	##	##	##
##	13	(1)	##	##	##	##	##	##	##
##	14	(1)	##	##	##	##	##	##	##
##	15	(1)	##	##	##	##	##	##	##
##	16	(1)	##	##	##	##	##	##	##
##	17	(1)	##	##	##	##	##	##	##
##	18	(1)	##	##	##	##	##	##	##
##	19	(1)	##	##	##	##	##	##	##
##	20	(1)	##	##	##	##	##	##	##
##	21	(1)	##	##	##	##	##	##	##
##	22	(1)	##	##	##	##	##	##	##
##	23	(1)	##	##	##	##	##	##	##
##	24	(1)	##	##	##	##	##	##	##
##	25	(1)	##	##	##	##	##	##	##
##	26	(1)	##	##	##	##	##	##	##
##	27	(1)	##	##	##	##	##	##	##
##	28	(1)	##	##	##	##	##	##	##
##	29	(1)	##	##	##	##	##	##	##
##	30	(1)	##	##	##	##	##	##	##
##	31	(1)	##	##	##	##	##	##	##
##	32	(1)	##	##	##	##	##	##	##
##	33	(1)	##	##	##	##	##	##	##
##	34	(1)	##	##	##	##	##	##	##
##	35	(1)	##	##	##	##	##	##	##
##	36	(1)	##	##	##	##	##	##	##
##	37	(1)	##	##	##	##	##	##	##
##	38	(1)	##	##	##	##	##	##	##
##	39	(1)	##	##	##	##	##	##	##
##	40	(1)	##	##	##	##	##	##	##
##	41	(1)	##	##	##	##	##	##	##
##	42	(1)	##	##	##	##	##	##	##
##	43	(1)	##	##	##	##	##	##	##
##	44	(1)	##	##	##	##	##	##	##
##	45	(1)	##	##	##	##	##	##	##
##	46	(1)	##	##	##	##	##	##	##
##			GK_Diving	GK_Kicking	GK_Handling	GK_Reflexes	Nationality_encoded		
##	1	(1)	##	##	##	##	##	##	##
##	2	(1)	##	##	##	##	##	##	##
##	3	(1)	##	##	##	##	##	##	##
##	4	(1)	##	##	##	##	##	##	##
##	5	(1)	##	##	##	##	##	##	##
##	6	(1)	##	##	##	##	##	##	##
##	7	(1)	##	##	##	##	##	##	##
##	8	(1)	##	##	##	##	##	##	##
##	9	(1)	##	##	##	##	##	##	##
##	10	(1)	##	##	##	##	##	##	##
##	11	(1)	##	##	##	##	##	##	##
##	12	(1)	##	##	##	##	##	##	##
##	13	(1)	##	##	##	##	##	##	##
##	14	(1)	##	##	##	##	##	##	##
##	15	(1)	##	##	##	##	##	##	##
##	16	(1)	##	##	##	##	##	##	##
##	17	(1)	##	##	##	##	##	##	##
##	18	(1)	##	##	##	##	##	##	##
##	19	(1)	##	##	##	##	##	##	##
##	20	(1)	##	##	##	##	##	##	##
##	21	(1)	##	##	##	##	##	##	##

##	22	(1)	**	**	**	**	**
##	23	(1)	**	**	**	**	**
##	24	(1)	**	**	**	**	**
##	25	(1)	**	**	**	**	**
##	26	(1)	**	**	**	**	**
##	27	(1)	**	**	**	**	**
##	28	(1)	**	**	**	**	**
##	29	(1)	**	**	**	**	**
##	30	(1)	**	**	**	**	**
##	31	(1)	**	**	**	**	**
##	32	(1)	**	**	**	**	**
##	33	(1)	**	**	**	**	**
##	34	(1)	**	**	**	**	**
##	35	(1)	**	**	**	**	**
##	36	(1)	**	**	**	**	**
##	37	(1)	**	**	**	**	**
##	38	(1)	**	**	**	**	**
##	39	(1)	**	**	**	**	**
##	40	(1)	**	**	**	**	**
##	41	(1)	**	**	**	**	**
##	42	(1)	**	**	**	**	**
##	43	(1)	**	**	**	**	**
##	44	(1)	**	**	**	**	**
##	45	(1)	**	**	**	**	**
##	46	(1)	**	**	**	**	**
##			Club_encoded	Preffered_Foot_encoded	Preffered_Position_encoded		
##	1	(1)	**	**	**		
##	2	(1)	**	**	**		
##	3	(1)	**	**	**		
##	4	(1)	**	**	**		
##	5	(1)	**	**	**		
##	6	(1)	**	**	**		
##	7	(1)	**	**	**		
##	8	(1)	**	**	**		
##	9	(1)	**	**	**		
##	10	(1)	**	**	**		
##	11	(1)	**	**	**		
##	12	(1)	**	**	**		
##	13	(1)	**	**	**		
##	14	(1)	**	**	**		
##	15	(1)	**	**	**		
##	16	(1)	**	**	**		
##	17	(1)	**	**	**		
##	18	(1)	**	**	**		
##	19	(1)	**	**	**		
##	20	(1)	**	**	**		
##	21	(1)	**	**	**		
##	22	(1)	**	**	**		
##	23	(1)	**	**	**		
##	24	(1)	**	**	**		
##	25	(1)	**	**	**		
##	26	(1)	**	**	**		
##	27	(1)	**	**	**		
##	28	(1)	**	**	**		
##	29	(1)	**	**	**		
##	30	(1)	**	**	**		
##	31	(1)	**	**	**		
##	32	(1)	**	**	**		
##	33	(1)	**	**	**		
##	34	(1)	**	**	**		
##	35	(1)	**	**	**		
##	36	(1)	**	**	**		
##	37	(1)	**	**	**		
##	38	(1)	**	**	**		
##	39	(1)	**	**	**		
##	40	(1)	**	**	**		
##	41	(1)	**	**	**		
##	42	(1)	**	**	**		
##	43	(1)	**	**	**		
##	44	(1)	**	**	**		
##	45	(1)	**	**	**		
##	46	(1)	**	**	**		
##			Work_Rate_encoded				


```
## 1 ( 1 ) " "
## 2 ( 1 ) " "
## 3 ( 1 ) " "
## 4 ( 1 ) " "
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## 7 ( 1 ) " "
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## 42 ( 1 ) " * "
## 43 ( 1 ) " * "
## 44 ( 1 ) " * "
## 45 ( 1 ) " * "
## 46 ( 1 ) " * "
```

```
fit.bs.summary <- summary(fit.bs)
fit.bs.summary$cp
```

```
## [1] 18234.30473 15484.08798 11176.01024 6867.18297 5558.41348 4472.00388
## [7] 3722.62801 2935.37777 2380.75077 1819.65950 1376.86198 1131.34033
## [13] 964.29745 833.10019 680.37833 593.88012 546.68197 499.10202
## [19] 456.79364 415.25800 374.72403 334.04022 296.91256 258.40823
## [25] 223.45571 186.91630 158.04029 136.51353 116.40004 103.82958
## [31] 93.77632 84.13763 75.90994 68.95863 63.48245 57.50740
## [37] 52.63902 48.47215 45.23005 44.31754 43.80642 43.60585
## [43] 43.89061 43.67975 45.04866 47.00000
```

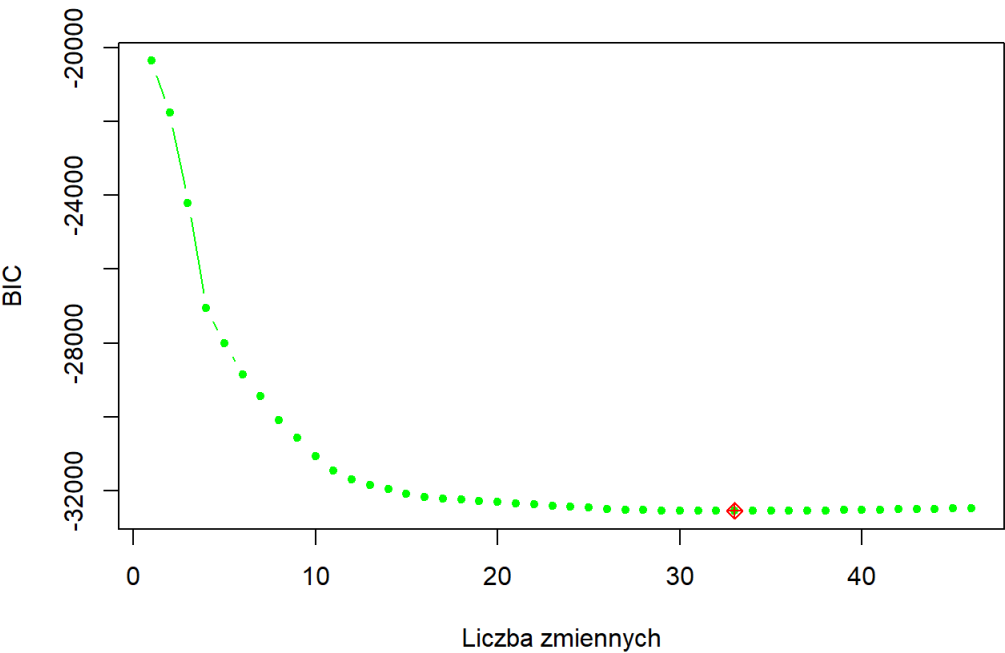
```
bic.min <- which.min(fit.bs.summary$bic)
bic.min
```

```
## [1] 33
```

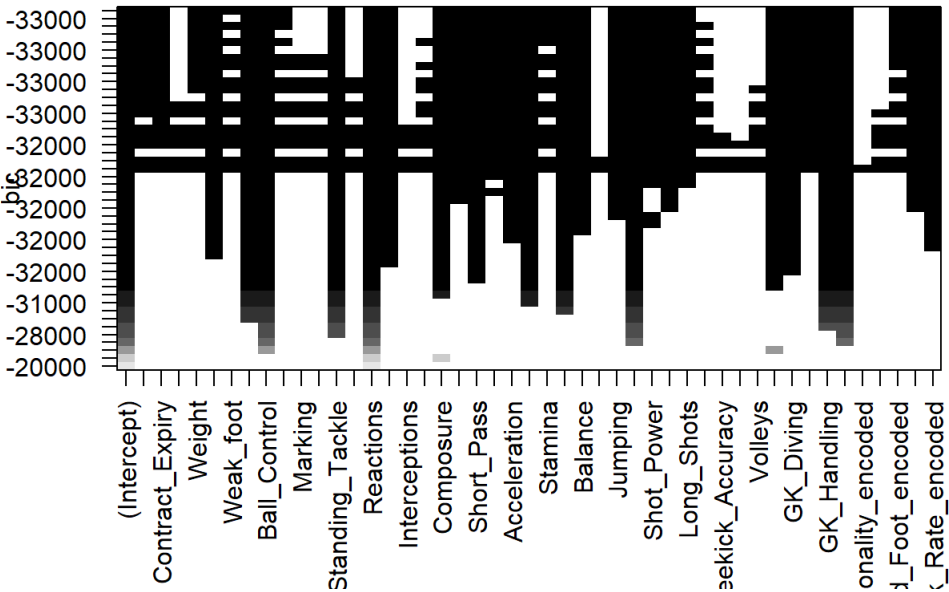
```
fit.bs.summary$bic[bic.min]
```

```
## [1] -32548.19
```

```
plot(fit.bs.summary$bic, xlab = "Liczba zmiennych", ylab = "BIC", col = "green",
     type = "b", pch = 20)
points(bic.min, fit.bs.summary$bic[bic.min], col = "red", pch = 9)
```



```
plot(fit.bs, scale = "bic")
```



```
coef(fit.bs, id = 6)
```

##	(Intercept)	Ball_Control	Standing_Tackle	Reactions	Heading
##	14.16940753	0.26906431	0.04794444	0.35353142	0.12813456
##	GK_Handling	GK_Reflexes			
##	0.16970715	0.16343786			

```
lmFit.many <- lm(Rating ~ Ball_Control + Standing_Tackle + Reactions + Heading + GK_Handling + GK_Reflexes)
summary(lmFit.many)
```

```
##
## Call:
## lm(formula = Rating ~ Ball_Control + Standing_Tackle + Reactions +
##     Heading + GK_Handling + GK_Reflexes)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -16.9149  -2.0465  -0.0067   2.0179  15.9982
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   14.168943   0.202836   69.85  <2e-16 ***
## Ball_Control    0.269066   0.002988   90.04  <2e-16 ***
## Standing_Tackle 0.047942   0.001373   34.93  <2e-16 ***
## Reactions       0.353534   0.003686   95.92  <2e-16 ***
## Heading        0.128140   0.002312   55.42  <2e-16 ***
## GK_Handling     0.169705   0.005760   29.46  <2e-16 ***
## GK_Reflexes     0.163443   0.005469   29.88  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.113 on 17581 degrees of freedom
## Multiple R-squared:  0.8069, Adjusted R-squared:  0.8068
## F-statistic: 1.224e+04 on 6 and 17581 DF, p-value: < 2.2e-16
```

```
lmFit.Standing_Tackle <- lm(Rating ~ Standing_Tackle, data = Fifa)
summary(lmFit.Standing_Tackle)
```

```
##
## Call:
## lm(formula = Rating ~ Standing_Tackle, data = Fifa)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -21.1305  -4.4235   0.0514   4.5159  29.1624
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   62.33242   0.12376   503.7  <2e-16 ***
## Standing_Tackle 0.08081   0.00237   34.1  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 6.86 on 17586 degrees of freedom
## Multiple R-squared:  0.06202, Adjusted R-squared:  0.06197
## F-statistic: 1163 on 1 and 17586 DF, p-value: < 2.2e-16
```

```
plot(Standing_Tackle, Rating)
abline(lmFit.Standing_Tackle)
```

